

# On $L^p$ -strong convergence of an averaging principle for non-Lipschitz slow-fast systems with Lévy noise

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## Abstract

We study  $L^p$ -strong convergence for coupled stochastic differential equations (SDEs) driven by Lévy noise with non-Lipschitz coefficients. Utilizing Khasminkii's time discretization technique, the Kunita's first inequality and Bihari's inequality, we show that the slow solution processes converge strongly in  $L^p$  to the solution of the corresponding averaged equation.

**Keywords.** Slow-fast systems, averaging principle, non-Lipschitz coefficients, Lévy noise.

**Mathematics subject classification.** 70K70, 60H10, 34K33

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## 1. Introduction

Let  $(\Omega, \mathcal{F}, P)$  be a complete probability space equipped with a natural filtration  $\{\mathcal{F}_t, t \geq 0\}$ , and  $\|\cdot\|$  denote the Euclidean norm as well as the matrix trace norm. Consider the following slow-fast stochastic differential equations with Lévy noise

$$\begin{cases} dX_t^\epsilon = a(X_t^\epsilon, Y_t^\epsilon) dt + b(X_t^\epsilon) dB_t^1 + \int_{\mathbb{Z}} \sigma(X_t^\epsilon, z) \tilde{N}_1(dt, dz), X_0^\epsilon = x_0, \\ dY_t^\epsilon = \frac{1}{\epsilon} f(X_t^\epsilon, Y_t^\epsilon) dt + \frac{1}{\sqrt{\epsilon}} g(X_t^\epsilon, Y_t^\epsilon) dB_t^2 + \int_{\mathbb{Z}} h(X_t^\epsilon, Y_t^\epsilon, z) \tilde{N}_2^\epsilon(dt, dz), Y_0^\epsilon = y_0, \end{cases} \quad (1.1)$$

for small parameter  $0 < \epsilon \ll 1$  and certain mappings  $a(u, v) : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n, b(u) : \mathbb{R}^n \rightarrow \mathbb{R}^{n \times d_1}, \sigma(u, z) : \mathbb{R}^n \times \mathbb{Z} \rightarrow \mathbb{R}^n, f(u, v) : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^m, g(u, v) : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^{m \times d_2}, h(u, v, z) : \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{Z} \rightarrow \mathbb{R}^m$ . The driving processes  $B_t^1$  and  $B_t^2$  are respectively  $d_1$ - and  $d_2$ -dimensional independent Brownian motions,  $\tilde{N}_1(dt, dz) = N_1(dt, dz) - \nu_1(dz) dt$ ,  $\tilde{N}_2^\epsilon(dt, dz) = N_2(dt, dz) - \frac{1}{\epsilon} \nu_2(dz) dt$  are compensated Poisson random measures, respectively, with intensity Lévy measure  $\nu_1$  and  $\nu_2$  on a measurable space  $(\mathbb{Z}, \mathcal{B}(\mathbb{Z}))$  of  $(-\infty, +\infty)$ , where  $\mathcal{B}(\mathbb{Z})$  denotes the Borel  $\sigma$ -algebra of  $\mathbb{Z}$ . Recently, [1] proposed a set of conditions on the non-Lipschitz coefficients (which is more general and much weaker than the usual Lipschitz conditions) ensuring the existence and uniqueness of solutions to the system (1.1).

The theory of averaging principle for SDEs was initiated by Khasminkii [2]. Since then, there are vast development devoted to this topic. For the slow-fast systems with noise, the

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averaging principle provides a powerful tool for approximating the slow-fast SDEs by reduced SDEs with slow component, see the recent work for systems with Gaussian white noise in [3, 4], the systems with Lévy-type perturbations in [5–7] and with non-Lipschitz coefficients in [8–10]. While most of the above mentioned references dealt mainly with square convergence, it is a natural and also very important question: Whether the validity of averaging principle to non-Lipschitz slow-fast systems still holds under higher order-convergence? Our aim in the present paper is to establish the convergence scheme in  $L^p$  sense (i.e., convergence in  $p$ -th moments) for the slow-fast system (1.1).

The rest of the paper is organized as follows. Section 2 presents some assumptions and preliminary results that are needed in the subsequent section. In Section 3, we prove that the slow component processes  $X_t^\epsilon$  converge to the limit process  $\bar{X}_t$  in the sense of  $p$ -th moments. Throughout the paper,  $C, C_p, C_T$  and  $C_{p,T}$  denote positive constants which may change from line to line, where  $C_p$  depends on  $p$ ,  $C_T$  depends on  $T$ , and  $C_{p,T}$  depends on  $p, T$ , etc..

## 2. Preliminaries

We assume the coefficients of system (1.1) fulfil the following

(A1) There exists a concave continuous nondecreasing function  $\kappa(\cdot)$ , such that

$$\begin{aligned} \|a(u_1, v_1) - a(u_2, v_2)\|^q + \|b(u_1) - b(u_2)\|^q + \int_{\mathbb{Z}} \|\sigma(u_1, z) - \sigma(u_2, z)\|^q \nu_1(dz) \\ \leq C \kappa(\|u_1 - u_2\|^q + \|v_1 - v_2\|^q), \end{aligned}$$

$$\begin{aligned} \|f(u_1, v_1) - f(u_2, v_2)\|^q + \|g(u_1, v_1) - g(u_2, v_2)\|^q \\ + \int_{\mathbb{Z}} \|h(u_1, v_1, z) - h(u_2, v_2, z)\|^q \nu_2(dz) \\ \leq C(\|u_1 - u_2\|^q + \|v_1 - v_2\|^q), \end{aligned}$$

for all  $u_j \in \mathbb{R}^n, v_j \in \mathbb{R}^m (j = 1, 2)$ ,  $q \geq 2$ , where  $\kappa(\cdot) : \mathbb{R}^+ \rightarrow \mathbb{R}^+$ ,  $\kappa(0) = 0$ ,  $\int_{0+} \frac{du}{\kappa(u)} = \infty$ . One can find a pair of positive constants  $\alpha_1$  and  $\alpha_2$  such that for  $\kappa(u) \leq \alpha_1 u + \alpha_2, \forall u \geq 0$ .

(A2) Assume that  $a$  is globally bounded,  $a, b, \sigma, f, g$  and  $h$  satisfy the linear growth conditions. Precisely, there exist constants  $\iota_1, \iota_2$  and  $\iota_3$  such that

$$\begin{aligned} \|a(u, v)\|^2 \vee \|b(u)\|^2 \vee \|f(u, v)\|^2 \vee \|g(u, v)\|^2 \leq \iota_1(1 + \|u\|^2 + \|v\|^2), \\ \int_{\mathbb{Z}} \|\sigma(u, z)\|^q \nu_1(dz) \leq \iota_2(1 + \|u\|^q), \quad \int_{\mathbb{Z}} \|h(u, v, z)\|^q \nu_2(dz) \leq \iota_3(1 + \|u\|^q + \|v\|^q), \end{aligned}$$

for all  $u \in \mathbb{R}^n, v \in \mathbb{R}^m, q \geq 2$ .

(A3) There exist constants  $c_0 > 0$  and  $c_j \in \mathbb{R}, j = 1, 2$ , which are all independent of  $(u_1, v_1, v_2)$ , such that

$$v_1^T f(u_1, v_1) \leq -c_0 \|v_1\|^2 + c_1, \quad (f(u_1, v_1) - f(u_1, v_2))^T (v_1 - v_2) \leq c_2 \|v_1 - v_2\|^2,$$

for all  $u_1 \in \mathbb{R}^n, v_1, v_2 \in \mathbb{R}^m$ , where T stands for the transpose.

(A4) There exists a constant  $\beta > 0$ , which is independent of  $(u, v)$ , such that

$$v^T g(u, v) g^T(u, v) v \geq \beta \|v\|^2.$$

**Remark 2.1.** According to Assumptions (A1)-(A2), the existence and uniqueness of solutions of system (1.1) can be obtained (see [1]). On the other hand, Assumptions (A3)-(A4) are two conditions which yield a unique invariant measure possessing exponentially mixing property to the fast variable motion (see [11]).

Now, we define the solution of the corresponding averaged equation:

$$d\bar{X}_t = \bar{a}(\bar{X}_t)dt + b(\bar{X}_t)dB_t + \int_{\mathbb{Z}} \sigma(\bar{X}_t, z) \tilde{N}_1(dt, dz), \quad \bar{X}_0 = x_0, \quad (2.1)$$

with the averaged coefficient  $\bar{a}(x) = \int_{\mathbb{R}^m} a(x, y) \mu^x(dy)$ , and  $\mu^x$  is the unique invariant measure for the transition semigroup of the frozen equation

$$dY_t = f(x, Y_t)dt + g(x, Y_t)d\bar{B}_t + \int_{\mathbb{Z}} h(x, Y_t, z) \bar{N}(dt, dz), \quad Y_0 = y_0. \quad (2.2)$$

for any fixed  $x \in \mathbb{R}^n$ .

**Remark 2.2.** For  $x \in \mathbb{R}^n, y \in \mathbb{R}^m$ , one has  $\|\mathbb{E}a(x, Y_t^{x,y}) - \bar{a}(x)\|^2 \leq Ce^{-\eta t}(1 + \|x\|^2 + \|y\|^2)$ , where  $\eta$  is a positive constant concerned with  $\iota_i (i = 1, 3)$  and  $c_j (j = 0, 1, 2)$ . This is due to the invariant property of  $\mu^x$  (see [12]).

**Lemma 2.3.** Assume that (A1)-(A4) are satisfied, then there exists a positive constant  $C_{p, \iota_i, c_j, T}$  such that for all  $\epsilon \in (0, 1), t \in [0, T], p \geq 1$ ,

$$\begin{aligned} \mathbb{E} \sup_{0 \leq t \leq T_0} \|X_t^\epsilon\|^{2p} &\leq C_{p, \iota_i, T}, \\ \sup_{0 \leq t \leq T_0} \mathbb{E} \|Y_t^\epsilon\|^{2p} &\leq C_{p, \iota_i, c_j, T}, \end{aligned}$$

where  $C_{p, \iota_i, c_j, T}$  is independent of  $\epsilon$ ,  $i = 1, 2, 3, j = 0, 1$ .

**Proof:** The techniques similar to those used in Lemma 3.1 and 3.2 in [13]. Here, the detailed proof is omitted for the sake of brevity.  $\square$

**Lemma 2.4.** Assume that (A1)-(A4) are satisfied, for all  $t \in [0, T], h \in (0, 1), p \geq 1$ , then there exists a positive constant  $C_{p, \iota_i, c_j, T}$  such that

$$\mathbb{E} \|X_{t+h}^\epsilon - X_t^\epsilon\|^{2p} \leq C_{p, \iota_i, c_j, T} h,$$

where  $C_{p, \iota_i, c_j, T}$  is independent of  $\epsilon$ ,  $i = 1, 2, 3, j = 0, 1$ .

**Proof:** Thanks to the Hölder inequality, Assumption (A2), the Burkholder's inequality and Burkholder-Davis-Gundy inequality, the following can be derived

$$\begin{aligned}
& \mathbb{E} \|X_{t+h}^\epsilon - X_t^\epsilon\|^{2p} \\
& \leq C_p \mathbb{E} \left\| \int_t^{t+h} a(X_s^\epsilon, Y_s^\epsilon) ds \right\|^{2p} + C_p \mathbb{E} \left\| \int_t^{t+h} b(X_s^\epsilon) dB_s^1 \right\|^{2p} + C_p \mathbb{E} \left\| \int_t^{t+h} \int_{\mathbb{Z}} \sigma(X_s^\epsilon, z) \tilde{N}_1(ds, dz) \right\|^{2p} \\
& \leq C_p h^{2p-1} \mathbb{E} \int_t^{t+h} \|a(X_s^\epsilon, Y_s^\epsilon)\|^{2p} ds + C_p \mathbb{E} \left[ \int_t^{t+h} \|b(X_s^\epsilon)\|^2 ds \right]^p \\
& \quad + C_p \mathbb{E} \left[ \int_t^{t+h} \int_{\mathbb{Z}} \|\sigma(X_s^\epsilon, z)\|^2 \nu_1(dz) ds \right]^p + C_p \mathbb{E} \left[ \int_t^{t+h} \int_{\mathbb{Z}} \|\sigma(X_s^\epsilon, z)\|^{2p} \nu_1(dz) ds \right] \\
& \leq C_{p, \iota_1} h^{2p-1} \int_t^{t+h} \mathbb{E} (1 + \|X_s^\epsilon\|^{2p} + \|Y_s^\epsilon\|^{2p}) ds + C_{p, \iota_1} h^{p-1} \int_t^{t+h} \mathbb{E} (1 + \|X_s^\epsilon\|^{2p}) ds \\
& \quad + C_{p, \iota_2} h^{p-1} \mathbb{E} \left[ \int_t^{t+h} (1 + \|X_s^\epsilon\|^2)^p ds \right] + C_{p, \iota_2} \mathbb{E} \left[ \int_t^{t+h} (1 + \|X_s^\epsilon\|^{2p}) ds \right] \\
& \leq C_{p, \iota_i, c_j, T} h.
\end{aligned}$$

This completes the proof of Lemma 2.4.  $\square$

**Remark 2.5.** We point out that the slow compolent contains the compensated Poisson random measure in Lemma 2.4. This makes an intrinsic difference between Gaussian noise and Lévy-type noise to regularity of  $X_t^\epsilon$ , which affecting the rate of convergence scale in practical engineering.

### 3. Stochastic averaging principle

In this part, we intend to prove that the slow component process  $X_t^\epsilon$  convergence strongly to the solution  $\bar{X}_t$  of the averaged equation.

**Theorem 3.1.** Assume that (A1)-(A4) are satisfied, then if we choose  $\delta = \epsilon \ln \epsilon^{-k}$ , for any  $T > 0$ ,  $p \geq 1$ , we obtain

$$\lim_{\epsilon \rightarrow 0} \mathbb{E} \sup_{0 \leq t \leq T} \|X_t^\epsilon - \bar{X}_t\|^{2p} = 0. \quad (3.1)$$

**Proof.** The proof is divided into three steps. In Step 1,  $\mathbb{E} \sup_{0 \leq t \leq T} \|X_s^\epsilon - \hat{X}_s^\epsilon\|^{2p}$  will be estimated, we then derive estimate of  $\mathbb{E} \sup_{0 \leq t \leq T} \|\hat{X}_t^\epsilon - \bar{X}_t\|^{2p}$  in Step 2. At last, through Step 1 and Step 2, the main result of Theorem 3.1 will be obtained.

**Step 1.** Following the idea inspired by Khasminskii [2], we introduce an auxiliary processes  $(\hat{X}_t^\epsilon, \hat{Y}_t^\epsilon) \in \mathbb{R}^n \times \mathbb{R}^m$ . We consider a partition of  $[0, T]$  into intervals of the identical length  $\delta$ , and the time interval  $\delta = \epsilon \ln \epsilon^{-k}$  ( $k > 0$ ) is selected small enough. We construct a process  $\hat{Y}_t^\epsilon$ , with initial datum  $\hat{Y}_t^\epsilon = y_0$ ,

$$\hat{Y}_t^\epsilon = Y_{k\delta}^\epsilon + \frac{1}{\epsilon} \int_{k\delta}^t f(X_{k\delta}^\epsilon, \hat{Y}_s^\epsilon) ds + \frac{1}{\sqrt{\epsilon}} \int_{k\delta}^t g(X_{k\delta}^\epsilon, \hat{Y}_s^\epsilon) dB_s^2 + \int_{k\delta}^t \int_{\mathbb{Z}} h(X_{k\delta}^\epsilon, \hat{Y}_s^\epsilon, z) \tilde{N}_2^\epsilon(ds, dz),$$

where  $X_{k\delta}^\epsilon$  and  $Y_{k\delta}^\epsilon$  are slow and fast solution processes at time  $k\delta$ , respectively for  $t \in [k\delta, \min\{(k+1)\delta, T\})$ ,  $k > 0$ . Define the  $\hat{X}_t^\epsilon$  processes by integral

$$\hat{X}_t^\epsilon = x_0 + \int_0^t a(X_{[s/\delta]\delta}^\epsilon, \hat{Y}_s^\epsilon) ds + \int_0^t b(X_s^\epsilon) dB_s^1 + \int_0^t \int_{\mathbb{Z}} \sigma(X_s^\epsilon, z) \tilde{N}_1(ds, dz),$$

for  $t \in [0, T]$ , where  $s(\delta) = \lfloor \frac{s}{\delta} \rfloor \delta$  is the nearest breakpoint proceeding  $s$ .

**Remark 3.2.** Using Itô's formula, Assumption (A1) and Lemma 2.4, we can get the corresponding result:  $\mathbb{E}\|Y_t^\epsilon - \hat{Y}_t^\epsilon\|^{2p} \leq C_{p, \iota_i, c_j, T} \frac{\delta^2}{\epsilon} e^{C'_{p, \iota_i, c_j, T} \frac{\delta}{\epsilon}}$ . Here, we omit the proof.

Next, it follows from definitions of  $X_t^\epsilon$  and  $\hat{X}_t^\epsilon$ , by Itô's formula, we get

$$\begin{aligned} & \mathbb{E} \sup_{0 \leq s \leq t} \|X_s^\epsilon - \hat{X}_s^\epsilon\|^{2p} \\ &= 2p \mathbb{E} \sup_{0 \leq s \leq t} \int_0^s \|X_r^\epsilon - \hat{X}_r^\epsilon\|^{2p-1} \|a(X_r^\epsilon, Y_r^\epsilon) - a(X_{[r/\delta]\delta}^\epsilon, \hat{Y}_r^\epsilon)\| dr \\ &\leq 2p \mathbb{E} \int_0^t \|X_s^\epsilon - \hat{X}_s^\epsilon\|^{2p-1} \|a(X_s^\epsilon, Y_s^\epsilon) - a(X_{[s/\delta]\delta}^\epsilon, \hat{Y}_s^\epsilon)\| ds \\ &\leq C_p \int_0^t \mathbb{E} \|X_s^\epsilon - \hat{X}_s^\epsilon\|^{2p} ds + C_p \int_0^t \mathbb{E} \|a(X_s^\epsilon, Y_s^\epsilon) - a(X_{[s/\delta]\delta}^\epsilon, \hat{Y}_s^\epsilon)\|^{2p} ds \quad (3.2) \\ &\leq C_p \int_0^t \mathbb{E} \sup_{0 \leq r \leq s} \|X_r^\epsilon - \hat{X}_r^\epsilon\|^{2p} ds + C_p \int_0^t \kappa(\mathbb{E} \|X_s^\epsilon - X_{[s/\delta]\delta}^\epsilon\|^{2p} + \mathbb{E} \|Y_s^\epsilon - \hat{Y}_s^\epsilon\|^{2p}) ds \\ &\leq C_p \int_0^t \mathbb{E} \sup_{0 \leq r \leq s} \|X_r^\epsilon - \hat{X}_r^\epsilon\|^{2p} ds + C_p \kappa(\delta + C_{p, \iota_i, c_j, T} \frac{\delta^2}{\epsilon} e^{C'_{p, \iota_i, c_j, T} \frac{\delta}{\epsilon}}) \\ &\leq C_p \kappa(\delta + C_{p, \iota_i, c_j, T} \frac{\delta^2}{\epsilon} e^{C'_{p, \iota_i, c_j, T} \frac{\delta}{\epsilon}}) e^{C_p t} := \varepsilon_0. \end{aligned}$$

The last term applies the Gronwall's inequality.

To proceed, we can conclude that the mapping  $\bar{a} : \mathbb{R}^n \rightarrow \mathbb{R}^n$  satisfies non-Lipschitz condition from equation (2.1). Actually,  $Y_t^{x_1, y}$  and  $Y_t^{x_2, y}$  are two solutions of (2.2), then for any  $x_1, x_2 \in \mathbb{R}^n, y \in \mathbb{R}^m$ , according to the Kunita's first inequality, one gets

$$\begin{aligned} \mathbb{E} \sup_{0 \leq s \leq t} \|Y_s^{x_1, y} - Y_s^{x_2, y}\|^{2p} &\leq C_p \mathbb{E} \sup_{0 \leq s \leq t} \left\| \int_0^s [f(x_1, Y_r^{x_1, y}) - f(x_2, Y_r^{x_2, y})] dr \right\|^{2p} \\ &\quad + C_p \mathbb{E} \sup_{0 \leq s \leq t} \left\| \int_0^s [g(x_1, Y_r^{x_1, y}) - g(x_2, Y_r^{x_2, y})] dB_r^2 \right\|^{2p} \\ &\quad + C_p \mathbb{E} \sup_{0 \leq s \leq t} \left\| \int_0^s \int_{\mathbb{Z}} [h(x_1, Y_r^{x_1, y}, z) - h(x_2, Y_r^{x_2, y}, z)] \tilde{N}_2(dr, dz) \right\|^{2p} \\ &\leq C_{p, T} \mathbb{E} \int_0^t (\|x_1 - x_2\|^{2p} + \|Y_s^{x_1, y} - Y_s^{x_2, y}\|^{2p}) ds \\ &\leq C_{p, T} \|x_1 - x_2\|^{2p} + C_{p, T} \int_0^t \mathbb{E} \sup_{0 \leq r \leq s} \|Y_r^{x_1, y} - Y_r^{x_2, y}\|^{2p} ds \end{aligned}$$

$$\leq C_{p,T} \|x_1 - x_2\|^{2p},$$

where  $C_{p,T}$  is independent of  $(x_1, x_2, y)$ . Moreover, using the property of the function  $k(\cdot)$  it is easy to get

$$\begin{aligned} \|\bar{a}(x_1) - \bar{a}(x_2)\|^{2p} &= \left\| \int_{\mathbb{R}^m} a(x_1, y) \mu^{x_1}(dy) - \int_{\mathbb{R}^m} a(x_2, y) \mu^{x_2}(dy) \right\|^{2p} \\ &\leq \mathbb{E} \|a(x_1, Y_t^{x_1, y}) - a(x_2, Y_t^{x_2, y})\|^{2p} \\ &\leq \kappa(\|x_1 - x_2\|^{2p}) + \kappa(\mathbb{E} \|Y_t^{x_1, y} - Y_t^{x_2, y}\|^{2p}) \\ &\leq C_{p,T} \kappa(\|x_1 - x_2\|^{2p}). \end{aligned}$$

This implies that the mapping  $\mathbb{R}^n \mapsto \mathbb{R}^n$  is non-Lipchitz continuous. By the linear growth condition of  $\bar{a}$ , we can derive that the existence and uniqueness of solution of equation (2.1).

**Step 2.** In this step, we will estimate the following estimate  $\mathbb{E} \sup_{0 \leq t \leq T} \|\hat{X}_t^\epsilon - \bar{X}_t\|^{2p}$ . It follows from the definitions of  $\bar{X}_t$  and  $\hat{X}_t^\epsilon$  that

$$\begin{aligned} &\mathbb{E} \sup_{0 \leq t \leq T} \|\hat{X}_t^\epsilon - \bar{X}_t\|^{2p} \\ &= \mathbb{E} \sup_{0 \leq t \leq T} \left\| \int_0^t \left( a(X_{[s/\delta]\delta}^\epsilon, \hat{Y}_s^\epsilon) - \bar{a}(X_s^\epsilon) \right) ds \right\|^{2p} \\ &\quad + \mathbb{E} \sup_{0 \leq t \leq T} \left\| \int_0^t \left( \bar{a}(X_s^\epsilon) - \bar{a}(\hat{X}_s^\epsilon) \right) ds \right\|^{2p} + \mathbb{E} \sup_{0 \leq t \leq T} \left\| \int_0^t \left( \bar{a}(\hat{X}_s^\epsilon) - \bar{a}(\bar{X}_s) \right) ds \right\|^{2p} \\ &\quad + \mathbb{E} \sup_{0 \leq t \leq T} \left\| \int_0^t \left( b(X_s^\epsilon) - b(\hat{X}_s^\epsilon) \right) dB_s^1 \right\|^{2p} + \mathbb{E} \sup_{0 \leq t \leq T} \left\| \int_0^t \left( b(\hat{X}_s^\epsilon) - b(\bar{X}_s) \right) dB_s^1 \right\|^{2p} \\ &\quad + \mathbb{E} \sup_{0 \leq t \leq T} \left\| \int_0^t \int_{\mathbb{Z}} \left( \sigma(X_s^\epsilon, z) - \sigma(\hat{X}_s^\epsilon, z) \right) \tilde{N}_1(ds, dz) \right\|^{2p} \\ &\quad + \mathbb{E} \sup_{0 \leq t \leq T} \left\| \int_0^t \int_{\mathbb{Z}} \left( \sigma(\hat{X}_s^\epsilon, z) - \sigma(\bar{X}_s, z) \right) \tilde{N}_1(ds, dz) \right\|^{2p} := \sum_{i=1}^7 \Theta_i(t). \end{aligned}$$

Using Hölder inequality, under the help of Burkholder-Davious-Gundy inequality, one obtains

$$\begin{aligned} \mathbb{E} \left( \sum_{i=2}^7 \sup_{0 \leq t \leq T} \|\Theta_i(t)\|^{2p} \right) &\leq C_{p,T} \int_0^T \kappa(\mathbb{E} \sup_{0 \leq s \leq t} \|X_s^\epsilon - \hat{X}_s^\epsilon\|^{2p}) dt \\ &\quad + C_{p,T} \int_0^T \kappa(\mathbb{E} \sup_{0 \leq s \leq t} \|\bar{X}_s - \hat{X}_s^\epsilon\|^{2p}) dt. \end{aligned} \quad (3.3)$$

Now to deal with  $\Theta_1(t)$ , such that for  $t \in [k_t\delta, (k_t+1)\delta \wedge T]$ ,  $k_t := \lfloor \frac{t}{\delta} \rfloor$ , we have representation in the form of

$$\Theta_1(t) = \sum_{i=0}^{k_t-1} \int_{i\delta}^{(i+1)\delta} (a(X_{i\delta}^\epsilon, \hat{Y}_s^\epsilon) - \bar{a}(X_{i\delta}^\epsilon)) ds + \sum_{i=0}^{k_t-1} \int_{i\delta}^{(i+1)\delta} (\bar{a}(X_{i\delta}^\epsilon) - \bar{a}(X_s^\epsilon)) ds$$

$$+ \int_{k_t \delta}^t (a(X_{k_t \delta}^\epsilon, \hat{Y}_s^\epsilon) - \bar{a}(X_s^\epsilon)) ds := \sum_{i=1}^3 \Theta_{1i}(t). \quad (3.4)$$

For  $\Theta_{12}(t)$ , utilizing Hölder inequality and non-Lipschitz continuity of  $\bar{a}$ , it follows that

$$\begin{aligned} \mathbb{E} \sup_{0 \leq t \leq T} \|\Theta_{12}(t)\|^{2p} &\leq \mathbb{E} \sup_{0 \leq t \leq T} \int_0^{k_t \delta} \|\bar{a}(X_{\lfloor \frac{s}{\delta} \rfloor \delta}^\epsilon) - \bar{a}(X_s^\epsilon)\|^{2p} ds \\ &\leq C_{p,T} \int_0^T \kappa(\mathbb{E} \|X_{\lfloor \frac{s}{\delta} \rfloor \delta}^\epsilon - X_s^\epsilon\|^{2p}) ds. \end{aligned} \quad (3.5)$$

We proceed next to the estimation of  $\Theta_{13}(t)$ , by linear growth bound of functions  $a$  and  $\bar{a}$ ,

$$\mathbb{E} \sup_{0 \leq t \leq T} \|\Theta_{13}(t)\|^{2p} \leq \mathbb{E} \sup_{0 \leq t \leq T} \int_{k_t \delta}^t \|a(X_{k_t \delta}^\epsilon, \hat{Y}_s^\epsilon) - \bar{a}(X_s^\epsilon)\|^{2p} ds \leq C_{p, \iota_i, c_j, T} \delta. \quad (3.6)$$

Again, as for  $\Theta_{11}(t)$  we can deduce that

$$\begin{aligned} \mathbb{E} \sup_{0 \leq t \leq T} \|\Theta_{11}(t)\|^{2p} &= \mathbb{E} \sup_{0 \leq t \leq T} \left\| \sum_{i=0}^{k_t-1} \int_{i\delta}^{(i+1)\delta} [a(X_{i\delta}^\epsilon, \hat{Y}_s^\epsilon) - \bar{a}(X_{i\delta}^\epsilon)] ds \right\|^{2p} \\ &\leq \mathbb{E} \sup_{0 \leq t \leq T} \left( \frac{t}{\delta} \right)^{2p-1} \sum_{i=0}^{k_t-1} \left\| \int_{i\delta}^{(i+1)\delta} [a(X_{i\delta}^\epsilon, \hat{Y}_s^\epsilon) - \bar{a}(X_{i\delta}^\epsilon)] ds \right\|^{2p} \\ &\leq \left( \frac{T}{\delta} \right)^{2p-1} \sum_{i=0}^{\lfloor \frac{T}{\delta} \rfloor - 1} \mathbb{E} \left\| \int_{i\delta}^{(i+1)\delta} [a(X_{i\delta}^\epsilon, \hat{Y}_s^\epsilon) - \bar{a}(X_{i\delta}^\epsilon)] ds \right\|^{2p} \\ &\leq \left( \frac{T}{\delta} \right)^{2p} \epsilon^{2p} \max_{0 \leq i \leq \lfloor \frac{T}{\delta} \rfloor - 1} \mathbb{E} \left\| \int_0^{\delta/\epsilon} [a(X_{i\delta}^\epsilon, \hat{Y}_{s\epsilon+i\delta}^\epsilon) - \bar{a}(X_{i\delta}^\epsilon)] ds \right\|^{2p} \\ &\leq \left( \frac{\epsilon}{\delta} \right)^{2p} \left( \frac{\delta}{\epsilon} \right)^{2p-2} \max_{0 \leq i \leq \lfloor \frac{T}{\delta} \rfloor - 1} \mathbb{E} \left\| \int_0^{\delta/\epsilon} [a(X_{i\delta}^\epsilon, \hat{Y}_{s\epsilon+i\delta}^\epsilon) - \bar{a}(X_{i\delta}^\epsilon)] ds \right\|^2 \\ &\leq \left( \frac{\epsilon}{\delta} \right)^2 \max_{0 \leq i \leq \lfloor \frac{T}{\delta} \rfloor - 1} \mathcal{I}_k^\epsilon \leq C \frac{\epsilon}{\delta}, \end{aligned} \quad (3.7)$$

where  $\mathcal{I}_i^\epsilon = \mathbb{E} \left\| \int_0^{\delta/\epsilon} [a(X_{i\delta}^\epsilon, \hat{Y}_{s\epsilon+i\delta}^\epsilon) - \bar{a}(X_{i\delta}^\epsilon)] ds \right\|^2 \leq C \frac{\delta}{\epsilon}$  (see [13], Lemma 4.1.). Thus, substituting (3.5) (3.6) and (3.7) into (3.4), we can estimate

$$\mathbb{E} \sup_{0 \leq t \leq T} \|\Theta_1(t)\|^{2p} \leq C_{p, \iota_i, T} (\delta + \frac{\epsilon}{\delta}) + \int_0^T \kappa(\mathbb{E} \|X_{\lfloor \frac{s}{\delta} \rfloor \delta}^\epsilon - X_s^\epsilon\|^{2p}) ds. \quad (3.8)$$

Taking (3.3) and (3.8) into account, it can be deduce that

$$\mathbb{E} \sup_{0 \leq t \leq T} \|\bar{X}_t - \hat{X}_t^\epsilon\|^{2p} \leq C_{p,T} \int_0^T \kappa(\mathbb{E} \sup_{0 \leq s \leq t} \|\bar{X}_s - \hat{X}_s^\epsilon\|^{2p}) dt + C_{p, \iota_i, T} (\delta + \frac{\epsilon}{\delta})$$

$$\begin{aligned}
& + \int_0^T \kappa(\mathbb{E} \sup_{0 \leq s \leq t} \|X_s^\epsilon - \hat{X}_s^\epsilon\|^{2p}) dt + \int_0^T \kappa(\mathbb{E} \|X_{\lfloor \frac{t}{\delta} \rfloor \delta} - X_t\|^{2p}) dt \\
& \leq C_{p,T} \int_0^T \left[ \kappa(\mathbb{E} \sup_{0 \leq s \leq t} \|\bar{X}_s - \hat{X}_s^\epsilon\|^{2p}) + \mathbb{E} \sup_{0 \leq s \leq t} \|\bar{X}_s - \hat{X}_s^\epsilon\|^{2p} \right] dt \\
& \quad + C_{p,\iota_i,T} \left( \kappa(\delta + \frac{\epsilon}{\delta} + \varepsilon_0) + (\delta + \frac{\epsilon}{\delta} + \varepsilon_0) \right) \\
& \leq C_{p,T} \int_0^T \tilde{\kappa}(\mathbb{E} \sup_{0 \leq s \leq t} \|\bar{X}_s - \hat{X}_s^\epsilon\|^{2p}) dt + C_{p,\iota_i,T} \tilde{\kappa}(\delta + \frac{\epsilon}{\delta} + \varepsilon_0),
\end{aligned}$$

where  $\tilde{\kappa}(u) = u + \kappa(u)$ , obviously,  $\tilde{\kappa}(0) = 0$ ,  $\int_{0+} \frac{du}{\tilde{\kappa}(u)} = \infty$ . Applying the Bihari's inequality to get,

$$\mathbb{E} \sup_{0 \leq t \leq T} \|\bar{X}_t - \hat{X}_t^\epsilon\|^{2p} \leq G^{-1} \left[ G(C_{p,\iota_i,T} \tilde{\kappa}(\delta + \frac{\epsilon}{\delta} + \varepsilon_0)) + C_{p,T} t \right],$$

where  $\delta = \epsilon \ln \epsilon^{-k}$  ( $0 < k < \frac{1}{2+C'_{p,\iota_i,c_j,T}}$ ),  $\lim_{\epsilon \rightarrow 0} \varepsilon_0 = \lim_{\epsilon \rightarrow 0} \kappa(\delta + C_{p,\iota_i,c_j,T} \frac{\delta^2}{\epsilon} e^{C'_{p,\iota_i,c_j,T} \frac{\delta}{\epsilon}}) e^{C_p t} = 0$ ,  $\tilde{\kappa}(\delta + \frac{\epsilon}{\delta} + \varepsilon_0) \rightarrow 0$ , recalling the condition  $\int_{0+} \frac{du}{\tilde{\kappa}(u)} = \infty$ , we have  $G(C_{p,\iota_i,T} \tilde{\kappa}(\delta + \frac{\epsilon}{\delta} + \varepsilon_0)) + C_{p,T} t \rightarrow -\infty$ , so we get  $G^{-1} [G(C_{p,\iota_i,T} \tilde{\kappa}(\delta + \frac{\epsilon}{\delta} + \varepsilon_0)) + C_{p,T} t] \rightarrow 0$ . Therefore we have

$$\lim_{\epsilon \rightarrow 0} \mathbb{E} \sup_{0 \leq t \leq T} \|\hat{X}_t^\epsilon - \bar{X}_t\|^{2p} = 0.$$

**Step 3.** In terms of the conclusions of Step 1 and Step 2, we obtain

$$\begin{aligned}
\mathbb{E} \sup_{0 \leq t \leq T} \|X_t^\epsilon - \bar{X}_t\|^{2p} &= \mathbb{E} \sup_{0 \leq t \leq T} \|X_t^\epsilon - \hat{X}_t^\epsilon + \hat{X}_t^\epsilon - \bar{X}_t\|^{2p} \\
&\leq C_p \mathbb{E} \sup_{0 \leq t \leq T} \|X_t^\epsilon - \hat{X}_t^\epsilon\|^{2p} + C_p \mathbb{E} \sup_{0 \leq t \leq T} \|\hat{X}_t^\epsilon - \bar{X}_t\|^{2p} \\
&\leq C_p \kappa(\delta + C_{p,\iota_i,c_j,T} \frac{\delta^2}{\epsilon} e^{C'_{p,\iota_i,c_j,T} \frac{\delta}{\epsilon}}) e^{C_p t}.
\end{aligned}$$

Consequently, taking  $\delta = \epsilon \ln \epsilon^{-k}$ , letting  $\epsilon \rightarrow 0$ , it is easy to see that Theorem 3.1 holds. The proof of Theorem 3.1 is completed.  $\square$

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